

Machine learning Model

- in this file we will be attempting to create a Machine Learning model that is able to identify and predict SYN flood from normal network communication.

Model Pseudocode

```
In [ ]: # BEGIN

# 1. Load dataset
#     - Separate features (X) and Labels (y)

# 2. Set up 5-fold cross-validation
#     - Use StratifiedKFold to maintain class balance

# 3. Define SVM model with:
#     - RBF kernel
#     - C = 10.0
#     - gamma = 'auto'
#     - class_weight = 'balanced'

# 4. For each fold:
#     a. Split data into training and validation sets
#     b. Train SVM on training set
#     c. Predict on validation set
#     d. Evaluate performance (e.g. accuracy)
#     e. Save results

# 5. Calculate and print average accuracy across all folds

# END
```

Step1: Select and Load the Dataset

```
In [2]: import pandas as pd

# Load the preprocessed and normalized dataset
df = pd.read_csv('D:\Coding Projects\Detection-of-SYN-Flood-Attacks-Using-Machine-L

# Separate features and Label
X = df.drop('Label', axis=1)
y = df['Label']

# Check the shapes
print(f'Features shape: {X.shape}')
print(f'Labels shape: {y.shape}')
```

Features shape: (9604, 13)

Labels shape: (9604,)

Step 2: Define SVM Model and Hyperparameters

```
In [ ]: from sklearn.svm import SVC

svm_model = SVC(
    C=10.0,
    kernel='rbf',
    gamma='auto',
    class_weight=None,
    probability=True,
    cache_size=500,
    tol=1e-4,
    shrinking=True,
    decision_function_shape='ovr',
    random_state=42
)
```

Step 3: Model Training with Cross-Validation + Resource Management

```
In [ ]: import time
import psutil
import os
from sklearn.metrics import accuracy_score

fold_accuracies = []
y_true = []
y_pred = []
y_scores = []

process = psutil.Process(os.getpid())

# Resource Monitoring Start
overall_start_time = time.time()
overall_start_ram = process.memory_info().rss / 1024 / 1024 # in MB
overall_start_cpu = psutil.cpu_percent(interval=1)

for fold in range(0, 5):
    print(f'\nTraining fold {fold}...')

    train_idx = df[df['Fold'] != fold].index
    val_idx = df[df['Fold'] == fold].index

    X_train, X_val = X.iloc[train_idx], X.iloc[val_idx]
    y_train, y_val = y.iloc[train_idx], y.iloc[val_idx]

    # Train model
    svm_model.fit(X_train, y_train)

    # Predict
    y_pred = svm_model.predict(X_val)
    y_scores = svm_model.decision_function(X_val)
```

```

# Store results
y_true.extend(y_val)
y_pred.extend(y_pred)
y_scores.extend(y_scores)

# Accuracy
acc = accuracy_score(y_val, y_pred)
fold accuracies.append(acc)
print(f'Fold {fold} Accuracy: {acc:.4f}')

# Resource Monitoring End
overall_end_time = time.time()
overall_end_ram = process.memory_info().rss / 1024 / 1024 # in MB
overall_end_cpu = psutil.cpu_percent(interval=1)

# Summary
print("\nOverall SVM Training Stats")
print(f"Total Training Time: {overall_end_time - overall_start_time:.2f} seconds")
print(f"Total RAM Usage Increase: {overall_end_ram - overall_start_ram:.2f} MB")
print(f"CPU Usage (at final check): {overall_end_cpu}%")

```

Training fold 0...
Fold 0 Accuracy: 0.9282

Training fold 1...
Fold 1 Accuracy: 0.9256

Training fold 2...
Fold 2 Accuracy: 0.9297

Training fold 3...
Fold 3 Accuracy: 0.9261

Training fold 4...
Fold 4 Accuracy: 0.9250

Overall SVM Training Stats
Total Training Time: 47.66 seconds
Total RAM Usage Increase: 0.04 MB
CPU Usage (at final check): 9.8%

Step 4: Model Evaluation

In [5]: `import numpy as np`

```

print("\nFinal SVM Cross-Validation Results:")
print(f'Accuracies from each fold: {fold accuracies}')
print(f'Average Accuracy: {np.mean(fold accuracies):.4f}')
print(f'Standard Deviation: {np.std(fold accuracies):.4f}')

```

Final SVM Cross-Validation Results:

Accuracies from each fold: [0.9281624154086413, 0.9255596043727226, 0.9297241020301926, 0.9260801665799063, 0.925]

Average Accuracy: 0.9269

Standard Deviation: 0.0018

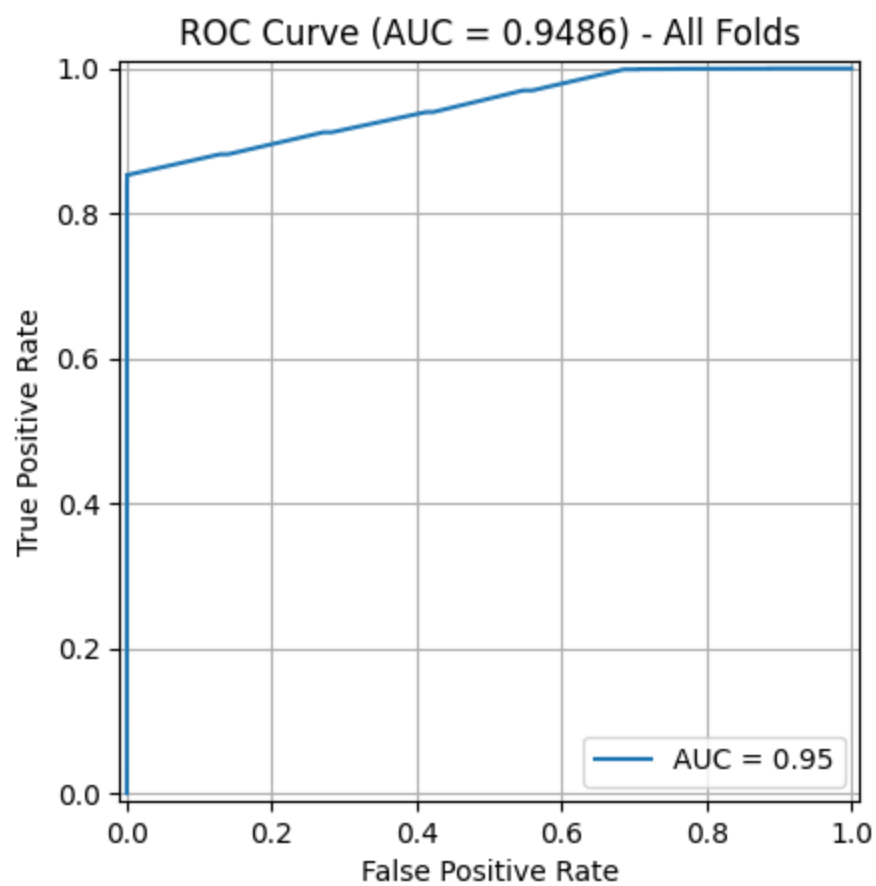
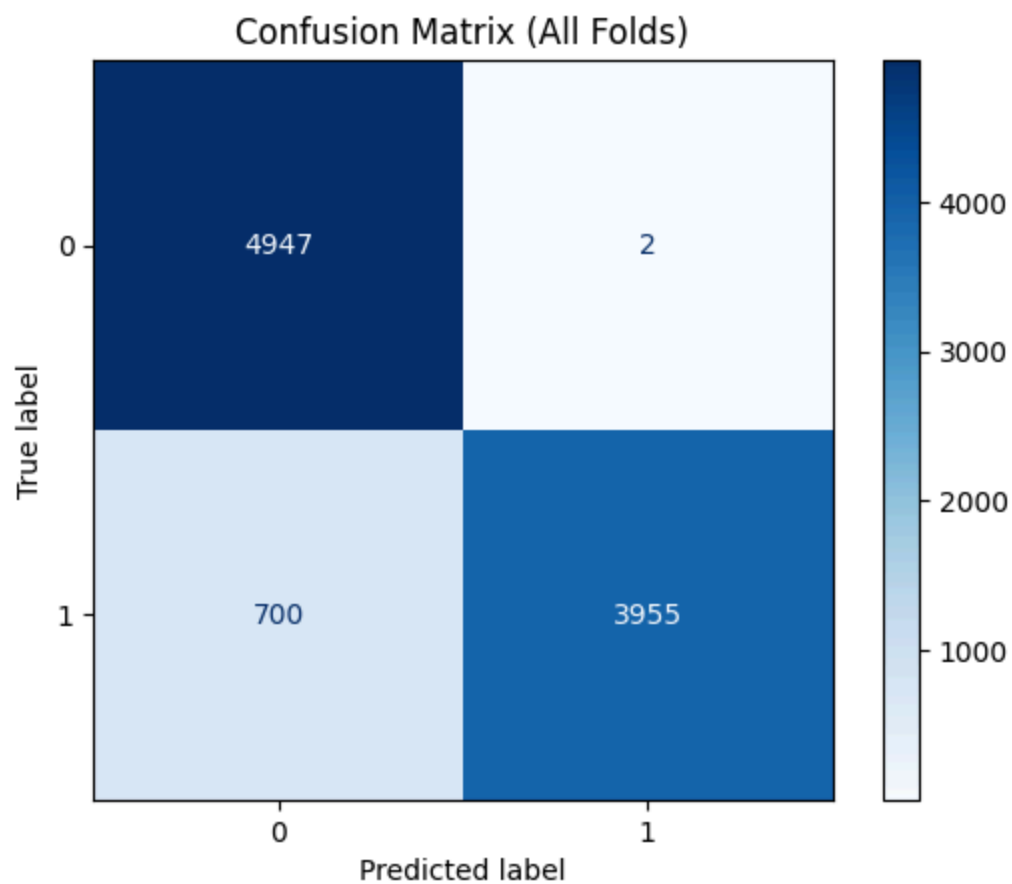
Step 5: Visual Evaluation

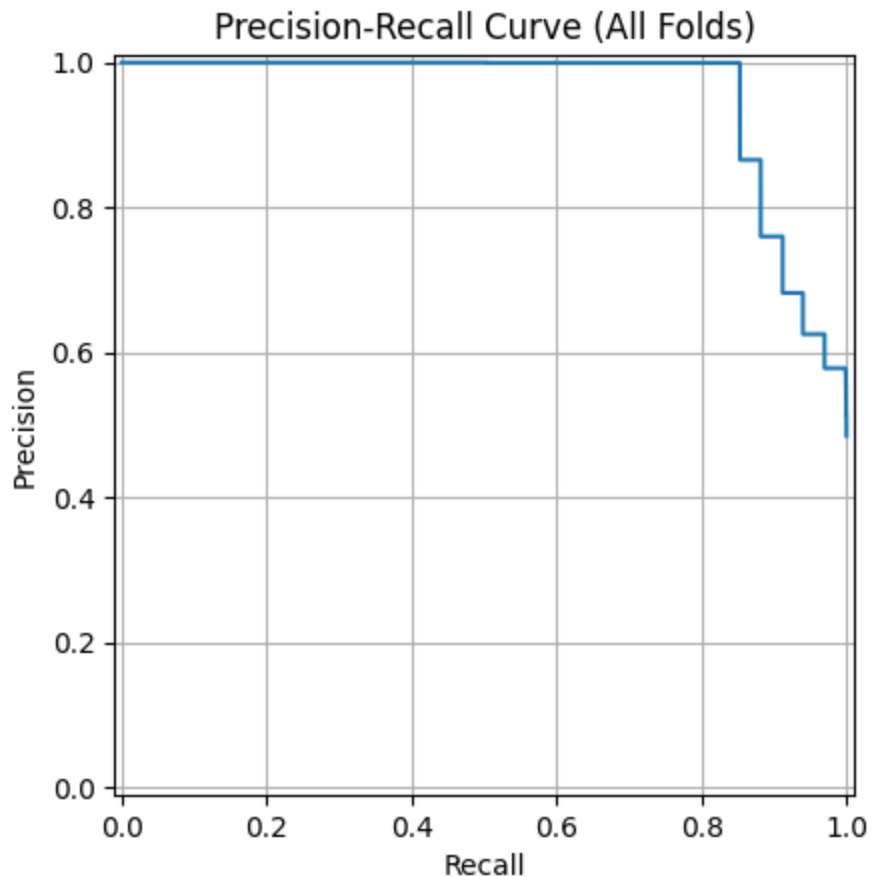
```
In [ ]: import matplotlib.pyplot as plt
        from sklearn.metrics import (
            confusion_matrix,
            ConfusionMatrixDisplay,
            roc_curve,
            auc,
            RocCurveDisplay,
            precision_recall_curve,
            PrecisionRecallDisplay,
        )

        # Confusion Matrix
        cm = confusion_matrix(y_true, y_pred)
        disp_cm = ConfusionMatrixDisplay(confusion_matrix=cm)
        disp_cm.plot(cmap='Blues')
        plt.title('Confusion Matrix (All Folds)')
        plt.grid(False)
        plt.show()

        # ROC Curve
        fpr, tpr, _ = roc_curve(y_true, y_scores)
        roc_auc = auc(fpr, tpr)
        RocCurveDisplay(fpr=fpr, tpr=tpr, roc_auc=roc_auc).plot()
        plt.title(f'ROC Curve (AUC = {roc_auc:.4f}) - All Folds')
        plt.grid(True)
        plt.show()

        # Precision-Recall Curve
        precision, recall, _ = precision_recall_curve(y_true, y_scores)
        PrecisionRecallDisplay(precision=precision, recall=recall).plot()
        plt.title('Precision-Recall Curve (All Folds)')
        plt.grid(True)
        plt.show()
```





saving the model as PDF

```
In [1]: import os  
os.getcwd()
```

```
Out[1]: 'd:\\Coding Projects\\Detection-of-SYN-Flood-Attacks-Using-Machine-Learning-and-Deep-Learning-Techniques-with-Feature-Base\\Amir Tavahin\\SVM.ipynb'
```

```
In [2]: !jupyter nbconvert --to webpdf "d:\\Coding Projects\\Detection-of-SYN-Flood-Attacks-Using-Machine-Learning-and-Deep-Learning-Techniques-with-Feature-Base\\Amir Tavahin\\SVM.ipynb"
```

```
[NbConvertApp] Converting notebook d:\\Coding Projects\\Detection-of-SYN-Flood-Attacks-Using-Machine-Learning-and-Deep-Learning-Techniques-with-Feature-Base\\Amir Tavahin\\SVM.ipynb to webpdf  
[NbConvertApp] WARNING | Alternative text is missing on 3 image(s).  
[NbConvertApp] Building PDF  
[NbConvertApp] PDF successfully created  
[NbConvertApp] Writing 200530 bytes to d:\\Coding Projects\\Detection-of-SYN-Flood-Attacks-Using-Machine-Learning-and-Deep-Learning-Techniques-with-Feature-Base\\Amir Tavahin\\SVM.pdf
```